

1. **Bias-variance.** Decompose the squared-error loss between a fixed (non-random) parameter  $\theta$  and its estimator  $\delta(X)$  into bias and variance terms. (Recall that the squared error is  $\mathbb{E}[(\delta(X) - \theta)^2]$ .)
2. **ROC Curves.** Consider the toy dataset in the table below;  $Y$  is the label,  $X_1, X_2$  are features, and we use the prediction function  $f(X_1, X_2)$ .

Table 1: Example dataset

$Y$	$f(X_1, X_2)$	$X_1$	$X_2$
0	-1	-1	0.5
1	-0.5	-1	0.75
0	0	-1	1
1	1	0.2	-0.3
1	0.25	-0.25	0
0	0.25	-0.05	-0.3

- (a) Draw the ROC curve for the prediction function  $f$  with respect to the label  $Y$ .
- (b) Is it possible to choose a (possibly randomized) decision threshold for  $f$ , such that the expected true positive rate is  $\frac{1}{3}$ , and the expected false positive rate is  $\frac{2}{3}$ ?

3. **LORD Procedure:**

You want to control the FDR with LORD at level  $\alpha$  and are currently at time step  $t = 5$ .

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**Algorithm 1** The LORD Procedure

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**Input** FDR level  $\alpha$ , non-increasing sequence  $\{\gamma_t\}_{t=1}^{\infty}$  such that  $\sum_{t=1}^{\infty} \gamma_t = 1$ ,  $W_0 = \alpha$

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- 1: Set  $\alpha_1 = \gamma_1 W_0$ .
  - 2: **for**  $t = 1, 2, \dots$ , **do**
  - 3:      $p$ -value  $P_t$  arrives.
  - 4:     **if**  $P_t \leq \alpha_t$  **then**
  - 5:         Reject  $P_t$ .
  - 6:     Update  $\alpha_{t+1} = \gamma_{t+1} W_0 + \alpha \sum_{j=1}^{\infty} \gamma_{t+1-\tau_j} 1\{\tau_j < t\}$ , where  $\tau_j$  is the time of the  $j$ 'th rejection.
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Set  $\gamma_t = 2^{-t}$ .

- (a) The only discovery you've made so far was at time step  $t = 1$ . How small must the 5'th  $p$ -value be in order for you to make a discovery at  $t = 5$ ?

(b) The only discovery you've made so far was at time step  $t = 2$ . How small must the 5'th  $p$ -value be in order for you to make a discovery at  $t = 5$ ?

(c) The only discovery you've made so far was at time step  $t = 3$ . How small must the 5'th  $p$ -value be in order for you to make a discovery at  $t = 5$ ?

(d) The only discovery you've made so far was at time step  $t = 4$ . How small must the 5'th  $p$ -value be in order for you to make a discovery at  $t = 5$ ?

(e) You're made discoveries at steps  $t = 1, 2, 3, 4$ . How small must the 5'th  $p$ -value be in order for you to make a discovery at  $t = 5$ ?

4. Alice has a bag with 3 red marbles, 2 blue marbles, and 1 green marble. Norman has a bag with 1 red marbles, 1 blue marble, and 4 green marbles. You observe two samples with replacement from either Alice or Norman, and want to figure out which is which. You want to have the highest TPR while keeping the FPR at  $\frac{1}{9}$ . What decision rule do you pick? What is the corresponding TPR? Assume that **Norman is the null** and **Alice is the alternative**.

To help you get started, the table below writes out the probabilities for all 9 outcomes:

	$P_A$	$P_N$
RR	1/4	1/36
RB	1/6	1/36
RG	1/12	1/9
BR	1/6	1/36
BB	1/9	1/36
BG	1/18	1/9
GR	1/12	1/9
GB	1/18	1/9
GG	1/36	4/9

### 5. Benjamini-Yekutieli procedure (Challenge Question)

Suppose you are testing  $n$  hypotheses and want to control the FDR at level  $\alpha$ . It turns out that Benjamini-Hochberg is only guaranteed to work when the hypotheses are *independent* or *positively correlated*. Construct an example with negatively correlated hypotheses where Benjamini-Hochberg fails.

**Remark:** The Benjamini-Yekutieli procedure, a generalization of Benjamini-Hochberg, controls the FDR regardless of independence assumptions, and therefore is guaranteed to work in all cases. It is shown below. The only difference from Benjamini-Hochberg is the  $c(n)$  function highlighted in red.

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**Algorithm 2** The Benjamini-Yekutieli Procedure

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**Input** FDR level  $\alpha$ , set of  $n$  p-values  $P_1, \dots, P_n$

- 1: Sort the p-values  $P_1, \dots, P_n$  in non-decreasing order  $P_{(1)} \leq P_{(2)} \leq \dots \leq P_{(n)}$
- 2: Find  $K = \max\{i \in \{1, \dots, n\} : P_{(i)} \leq \frac{\alpha}{n \cdot c(n)} i\}$ , where

$$c(n) = \begin{cases} 1 & \text{tests are independent or positively correlated (this is just B-H)} \\ \sum_{j=1}^n \frac{1}{j} & \text{tests are dependent or negatively correlated} \end{cases} \quad (1)$$

- 3: Reject the null hypotheses (declare discoveries) corresponding to  $P_{(1)}, \dots, P_{(K)}$
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